- ¹ Title: Tree height and leaf drought tolerance traits shape growth responses across droughts in a temperate
- 2 broadleaf forest
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22 Summary

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- As climate change is driving increased drought frequency and severity in many forested regions around the world, mechanistic understanding of the factors conferring drought resistance in trees is increasingly important. The dendrochronological record provides a window through which we can understand how tree size and species' traits shape tree growth responses during droughts.
- We analyzed tree-ring records for twelve species that comprise 97% of the woody productivity of the 25.6-ha ForestGEO plot in a broadleaf deciduous forest of northern Virginia (USA) to test hypotheses on how tree height, microenvironment characteristics, and species' traits shaped drought responses across the three strongest regional droughts over a 60-year period (1950 2009).
 - Individual-level drought resistance decreased with tree height, which was strongly correlated with
 exposure to higher evaporative demand and solar radiation. The potentially greater rooting volume
 of larger trees did not confer an advantage in sites with low topographic wetness index. Resistance
 was greater among species whose leaves experienced less shrinkage upon desiccation and lost turgor
 (wilted) at more negative water potentials.
- We conclude that tree height and leaf drought tolerance traits influence growth responses during
 drought, as recorded in the tree-ring record spanning historical droughts. Thus, these factors can be
 useful for predicting future drought responses under climate change.
- Key words: annual growth; crown exposure; drought; Forest Global Earth Observatory (ForestGEO); leaf drought tolerance traits; temperate broadleaf deciduous forest; tree height; tree-ring

Introduction

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Forests play a critical global role in climate regulation (Bonan, 2008), yet there remains enormous
   uncertainty as to how the forest-dominated terrestrial carbon sink will respond to climate change
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   (Friedlingstein et al., 2006). An important aspect of this uncertainty lies with physiological responses of
   trees to drought (Kennedy et al., 2019). In many forested regions around the world, the risk of severe
   drought is increasing (Trenberth et al., 2014; Dai et al., 2018), often despite increasing precipitation
   (Intergovernmental Panel on Climate Change, 2015; Cook et al., 2015). Droughts, intensified by climate
   change, have been affecting forests worldwide and are expected to continue as one of the most important
   drivers of forest change in the future (Allen et al., 2010, 2015). Understanding forest responses to drought
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   requires elucidation of how tree size, microenvironment, and species' traits jointly influence individual-level
   drought resistance, and the extent to which their influence is consistent across droughts. Because the
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   resistance and resilience of growth to drought is linked to trees' probability of surviving drought (DeSoto
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   et al., 2020; Liu et al., 2019), understanding growth responses can also help elucidate which trees are most
   vulnerable to drought-induced mortality. However, it has proven difficult to resolve the many factors
   affecting tree growth during drought with available forest census data, which only rarely captures extreme
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   drought, and with tree-ring records, which capture multiple droughts but usually only sample a subset of a
   forest community, typically focusing on a single species or the largest individuals.
   Many studies have shown that within and across species, large trees tend to be more affected by drought.
   Greater growth reductions for larger trees were first shown on a global scale by Bennett et al. (2015), and
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   subsequent studies have reinforced this finding (e.g., Hacket-Pain et al., 2016). It has yet to be resolved
   which of several potential underlying mechanisms most strongly shape these trends in drought response.
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   First, tree height itself may be a primary driver. Taller trees face the biophysical challenge of lifting water
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   greater distances against the effects of gravity and friction (McDowell et al., 2011; McDowell and Allen,
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   2015; Ryan et al., 2006; Couvreur et al., 2018). Vertical gradients in stem and leaf traits-including smaller
   and thicker leaves (higher leaf mass per area, LMA), greater resistance to hydraulic dysfunction (i.e., more
   negative water potential at 50% loss of hydraulic conductivity, more negative P50), and lower hydraulic
   conductivity at greater heights (Couvreur et al., 2018; Koike et al., 2001; McDowell et al., 2011)—enable
   trees to become tall (Couvreur et al., 2018). Greater stem capacitance (i.e., water storage capacity) of
   larger trees may also confer resistance to transient droughts [Phillips et al. (2003)*; Scholz et al. (2011)].
   Indeed, tall trees require xylem of greater hydraulic efficiency, such that xylem conduit diameters are wider
   in the basal portions of taller trees, both within and across species (Olson et al., 2018; Liu et al., 2019),
   and throughout the conductive systems of angiosperms (Zach et al., 2010; Olson et al., 2014, 2018). Wider
   xylem conduits plausibly make large trees more vulnerable to embolism during drought (Olson et al.,
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   2018), and traits conducive to efficient water transport may also lead to poor ability to recover from or
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   re-route water around embolisms (Roskilly et al., 2019).
   Larger trees may also have lower drought resistance because of microenvironmental and ecological factors.
   Their crowns tend to occupy more exposed canopy positions, which are associated with higher evaporative
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   demand (Kunert et al., 2017). Subcanopy trees tend to fare better specifically due to the benefits of a
   buffered environment (Pretzsch et al., 2018). Counteracting the liabilities associated with tall height, large
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   trees tend to have larger root systems (Enquist and Niklas, 2002), potentially mitigating some of the
   biophysical challenges they face by allowing greater access to water. Larger root systems—if they grant
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access to deeper water sources—would be particularly advantageous in drier microenvironments (e.g.,

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drought can be modified by species' traits and their distribution across size classes (Meakem et al., 2018;
    Liu et al., 2019). Understanding the mechanisms driving the greater relative growth reductions of larger
    trees during drought requires sorting out the interactive effects of height and associated exposure, root
    water access, and species' traits.
    Debates have also arisen regarding the traits influencing tree growth responses to drought. Studies within
    temperate broadleaf forests have observed ring-porous species showing higher drought tolerance than
    diffuse-porous species (Friedrichs et al., 2009; Elliott et al., 2015; Kannenberg et al., 2019), but this
    distinction would not hold in the global context (Wheeler et al., 2007; Olson et al., 2020) and does not
    resolve differences among the many species within each category. Commonly-measured traits including
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    wood density and leaf mass per area (LMA) have been linked to drought responses within some temperate
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    deciduous forests (Abrams, 1990; Guerfel et al., 2009; Hoffmann et al., 2011; Martin-Benito and Pederson,
    2015) and across forests worldwide (Greenwood et al., 2017). However, in other cases these traits could not
    explain drought tolerance (e.g., in a tropical rainforest; Maréchaux et al., 2019), or the direction of
    response was not always consistent. For instance, higher wood density has been associated with greater
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    drought resistance at a global scale (Greenwood et al., 2017), but correlated negatively with tree
    performance during drought in a broadleaf deciduous forest in the southeastern United States (Hoffmann
    et al., 2011). Thus, the perceived influence of these traits on drought resistance may actually reflect
    indirect correlations with other traits that more directly drive drought responses (Hoffmann et al., 2011).
101
    In contrast, hydraulic traits have direct physiological linkages to tree growth and mortality responses to
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    drought. For instance, water potentials at which percent the loss of conductivity surpasses a certain
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    threshold (e.g., P50 and P88, representing 50 and 88% loss of conductivity, respectively) and hydraulic
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    safety margin (i.e., difference between typical minimum water potentials and P50 or P88) correlate with
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    drought performance across global forests (Anderegg et al., 2016). However, these are time-consuming to
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    measure and therefore infeasible for predicting or modeling drought responses in highly diverse forests
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    (e.q., in the tropics). More easily-measurable leaf drought tolerance traits that have direct linkage to plant
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    hydraulic function can explain variation in plant distribution and function (Medeiros et al., 2019). These
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    include leaf area shrinkage upon desiccation (PLA_{dry}; Scoffoni et al., 2014) and the leaf water potential at
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    turgor loss point (\pi_{tlp}), i.e., the water potential at which leaf wilting occurs (Bartlett et al., 2016a; Zhu
111
    et al., 2018). Both traits correlate with hydraulic vulnerability and drought tolerance as part of unified
    plant hydraulic systems (Scoffoni et al., 2014; Bartlett et al., 2016a; Zhu et al., 2018; Farrell et al., 2017).
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    The abilities of both PLA_{dry} and \pi_{tlp} to explain tree drought resistance remains untested.
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    Here, we examine how tree height, microenvironment characteristics, and species' traits collectively shape
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    drought resistance, defined as the ratio of annual growth in a drought year to that which would be
    expected in the absence of drought based on previous years' growth. We test a series of hypotheses and
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    associated specific predictions (Table 1) based on the combination of tree-ring records from the three
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    strongest droughts over a 60-year period (1950 - 2009), species trait measurements, and census and
    microenvironmental data from a large forest dynamics plot in Virginia, USA. First, we focus on how tree
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    size, alone and in its interaction with microenvironmental gradients, influences drought resistance. We
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    examine the contemporary relationship between tree height and microenvironment, including growing
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    season meteorological conditions and crown exposure. We then test whether, consistent with most forests
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    globally, larger-diameter, taller trees tend to have lower drought resistance in this forest, which is in a
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hilltops, as compared to valleys and streambeds) during drought. Finally, tree size-related responses to

region (eastern North America) represented by only two studies in the global review of Bennett et al. (2015). We also test for an influence of potential access to available soil water, which should be greater for larger trees in dry but not in perpetually wet microsites. Finally, we focus on the role of species' traits, testing the hypothesis that species' traits—particularly leaf leaf drought tolerance traits—predict drought resistance. We test predictions that drought resistance is higher in ring-porous than semi-ring and diffuse-porous species and that it is correlated with wood density—either positively (Greenwood et al., 2017) or negatively (Hoffmann et al., 2011) and positively correlated with LMA. We further test predictions that species with low PLA_{dry} have higher drought resistance, and that species whose leaves lose turgor lower water potentials (more negative π_{tlp}) have higher resistance.

134 Materials and Methods

135 Study site and microclimate

Research was conducted at the 25.6-ha ForestGEO (Forest Global Earth Observatory) study plot at the Smithsonian Conservation Biology Institute (SCBI) in Virginia, USA (38°53'36.6"N, 78°08'43.4"W; Fig. S1) (Bourg et al., 2013; Anderson-Teixeira et al., 2015a). SCBI is located in the central Appalachian Mountains near the northern boundary of Shenandoah National Park. Elevations range from 273 to 338 m above sea level with a topographic relief of 65m (Bourg et al., 2013). Climate is humid temperate, with mean annual temperature of 12.7°C and precipitation of 1005 mm yr⁻¹ during our study period (1960-2009; source: CRU TS v.4.01; Harris et al., 2014). Dominant tree taxa within this secondary forest include Liriodendron tulipifera, oaks (Quercus spp.), and hickories (Carya spp.; Table 2).

We identified the three largest droughts within the time period 1950-2009, defining drought (Slette et al.,

 $_{144}$ $Identifying\ drought\ years$

Data collection and preparation

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2019) as events with anomalously dry peak growing season climatic conditions. Specifically, we used the 146 metric of Palmer Drought Severity Index (PDSI) during May-August (MJJA; Table S1), which were identified by Helcoski et al. (2019) as the months of the current year to which annual tree growth was most 148 sensitive at this site. PDSI divisional data for Northern Virginia were obtained from NOAA 149 (https://www7.ncdc.noaa.gov/CDO/CDODivisionalSelect.jsp) in December 2017. Based on this, we identified the three strongest droughts during the study period (Figs. 1, S1; Table S1). 151 The droughts differed in intensity and antecedent moisture conditions (Fig. S1, Table S1). The 1966 152 drought was preceded by two years of moderate drought during the growing season and severe to extreme 153 drought starting the previous fall. In August 1966, PDSI reached its lowest monthly value (-4.82) of the 154 three droughts. The 1977 drought was the least intense throughout the growing season, and it was 155 preceded by 2.5 years of near-normal conditions, making it the mildest of the three droughts. The 1999 156 drought was preceded by wetter than average conditions until the previous June, but PDSI plummeted 157 below -3.0 in October 1998 and remained below this threshold through August 1999. 158

Within or just outside the ForestGEO plot, we collected data on a suite of variables including tree heights, microenvironment characteristics, and species traits (Table 3). The SCBI ForestGEO plot was censused in 2008, 2013, and 2018 following standard ForestGEO protocols, whereby all free-standing woody stems \geq 1cm diameter at breast height (DBH) were mapped, tagged, measured at DBH, and identified to species

DBH and data for all stems ≥ 10cm to analyze functional trait composition relative to tree height (all analyses described below). Census data are available through the ForestGEO data portal 166 (www.forestgeo.si.edu). 167 We analyzed tree-ring data (xylem growth increment) from 571 trees representing the twelve dominant 168 species (Table 2; Fig. S2). Selected species were those with the greatest contributions to woody aboveground net primary productivity $(ANPP_{stem})$ and together comprised 97% of study plot $ANPP_{stem}$ 170 between 2008 and 2013 (Helcoski et al., 2019). Cores (one per tree) were collected within the ForestGEO 171 plot at breast height (1.3m) in 2010-2011 or 2016-2017. In 2010-2011, cores were collected from randomly selected live trees of each species that had at least 30 individuals ≥ 10 cm DBH (Bourg et al., 2013). 173 Annual tree mortality censuses were initiated in 2014 (Gonzalez-Akre et al., 2016), and in 2016-2017, cores 174 were collected from all trees found to have died since the previous year's census. We note that drought was probably not a cause of mortality for these trees, as monthly May-Aug PDSI did not drop below -1.75 in 176 these years or the three years prior (2013-2017), and that trees cored dead displayed similar climate 177 sensitivity to trees cored live (Helcoski et al., 2019). Cores were sanded, measured, and crossdated using standard procedures, as detailed in (Helcoski et al., 2019). The resulting chronologies (Fig. 1a) were 179 published in Zenodo (DOI: 10.5281/zenodo.2649302) in association with Helcoski et al. (2019). 180 For each cored tree, we combined tree-ring records and allometric equations of bark thickness to 181 reconstruct DBH for the years 1950-2009. Prior DBH was estimated using the following equation:

(Condit, 1998). From these census data, we used measurements of DBH from 2008 to calculate historical

$$DBH_Y = DBH_{2008} - 2 * \left[r_{bark,2008} - r_{bark,Y} + \sum_{year=Y}^{2008} r_{ring,Y} \right]$$

Here, Y denotes the year of interest, r_{ring} denotes ring width derived from cores, and r_{bark} denotes bark thickness. Bark thickness was estimated from species-specific allometries based on the bark thickness data

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from the site (Anderson-Teixeira et al., 2015b). Specifically, we used linear regression on log-transformed data to relate r_{bark} to diameter inside bark from 2008 data (Table S2), which were then used to determine 186 r_{bark} in the DBH reconstruction. 187 Tree heights (H) were measured by several researchers for a variety of purposes between 2012 and 2019 188 (n=1,518 trees). Methods included direct measurements using a collapsible measurement rod on small trees (NEON, 2018) or a tape measure on recently fallen trees (this study); geometric calculations using clinometer and tape measure (Stovall et al., 2018a) or digital rangefinders (Anderson-Teixeira et al., 2015b; 191 NEON, 2018); and ground-based LiDAR (Stovall et al., 2018b). Rangefinders used either the tangent method (Impulse 200LR, TruPulse 360R) or the sine method (Nikon ForestryPro) for calculating heights. 193 Both methods are associated with some error (Larjavaara and Muller-Landau, 2013), but in this instance 194 there was no clear advantage of one or the other. Measurements from the National Ecological Observatory Network (NEON) were collected near the ForestGEO plot following standard NEON protocol, whereby vegetation of short stature was measured with a collapsible measurement rod, and taller trees with a 197 rangefinder (NEON, 2018). Species-specific height allometries were developed (Table S3) using log-log regression $(ln[H] \sim ln[DBH])$. For species with insufficient height data to create reliable species-specific 199 allometries (n=2, JUNI and FRAM), heights were calculated from an equation developed by combining the 200 height measurements across all species. We then used these allometries to estimate H for each drought

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year, Y, based on reconstructed DBH_Y. The distribution of H across drought years is shown in Fig. S3.
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    To characterize how environmental conditions vary with height, data were obtained from the NEON tower
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    located <1km from the study area via the neonUtilities package (Lunch et al., 2020). We used wind speed,
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    relative humidity, and air temperature data, all measured over a vertical profile spanning heights from 7.2
205
    m to above the top of the tree canopy (31.0 or 51.8m, depending on censor), for the years 2016-2018
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    (NEON, 2018). After filtering for missing and outlier values, we determined the daily minima and maxima.
    which we then aggregated at the monthly scale.
208
    Crown position—a categorical variable classifying trees based on exposure to sunlight—was recorded for all
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    cored trees that remained standing during the growing season of 2018 following the protocol of Jennings
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    et al. (1999). Trees were classified as follows: dominant trees were defined as those with crowns above the
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    general level of the canopy, co-dominant trees as those with crowns within the canopy; intermediate
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    trees as those with crowns below the canopy level, but illuminated from above; and suppressed as those
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    below the canopy and receiving minimal direct illumination from above.
    Topographic wetness index (TWI), used here as a metric of long-term mean moisture availability, was
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    calculated using the dynatopmodel package in R (Fig. S2) (Metcalfe et al., 2018). Originally developed by
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    Beven and Kirkby (1979), TWI was part of a hydrological run-off model and has since been used for a
    number of purposes in hydrology and ecology (Sørensen et al., 2006). TWI calculation depends on an input
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    of a digital elevation model (DEM; ~3.7 m resolution from the elevatr package (Hollister, 2018)), and from
219
    this yields a quantitative assessment defined by how "wet" an area is, based on areas where run-off is more
    likely. From our observations in the plot, TWI performed better at categorizing wet areas than the
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    Euclidean distance from the stream.
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    Species' trait data were collected in August 2018 (Tables 2-3; Fig. S4). We sampled small, sun-exposed
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    branches up to eight meters above the ground from three individuals of each species in and around the
    ForestGEO plot. Sampled branches were re-cut under water at least two nodes above the original cut and
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    re-hydrated overnight in covered buckets under opaque plastic bags before measurements were taken.
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    Rehydrated leaves taken towards the apical end of the branch (n=3 per individual: small, medium, and
227
    large) were scanned, weighed, dried at 60^{\circ} C for \geq 48 hours, and then re-scanned and weighed. Leaf area
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    was calculated from scanned images using the LeafArea R package (Katabuchi, 2019). LMA was
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    calculated as the ratio of leaf dry mass to fresh area. PLA_{dry} was calculated as the percent loss of area
230
    between fresh and dry leaves. Wood density was calculated for ~1cm diameter stem samples (bark and pith
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    removed) as the ratio of dry weight to fresh volume, which was estimated using Archimedes' displacement.
232
    We used the rapid determination method of Bartlett et al. (2012) to estimate osmotic potential at turgor
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    loss point (\pi_{tlp}). Briefly, two 4 mm diameter leaf discs were cut from each leaf, tightly wrapped in foil,
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    submerged in liquid nitrogen, perforated 10-15 times with a dissection needle, and then measured using a
235
    vapour pressure osmometer (VAPRO 5520, Wescor, Logan, UT, USA). Osmotic potential (\pi_{osm}) given by
    the osmometer was used to estimate (\pi_{tlp}) using the equation \pi_{tlp} = 0.832\pi_{osm}^{-0.631} (Bartlett et al., 2012).
237
    Statistical Analysis
238
    For each drought year, we calculated a metric drought resistance (Rt) as the ratio of basal area increment
239
    (BAI; i.e., change in cross-sectional area) during the drought year to the mean BAI over the five years
    preceding the drought (Lloret et al., 2011). Thus, Rt values <1 and >1 indicate growth reductions and
241
    increases, respectively. Because the Rt metric could be biased by directional pre-drought growth trends, we
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also tried an intervention time series analysis (ARIMA, (Hyndman et al., 2020)) that predicted mean
243
    drought-year growth based on trends over the preceding ten years and used this value in place of the
    five-year mean in calculations of resistance (Rt_{ARIMA} = observed BAI/ predicted BAI). The two metrics
245
    were strongly correlated (Fig. S5). Visual review of the individual tree-ring sequences with the largest
246
    discrepancies between these metrics revealed that Rt was less prone to unreasonable estimates than
    Rt_{ARIMA}, so we selected Rt as our focal metric, presenting parallel results for Rt_{ARIMA} in the
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    Supplementary Info. In this study we focus exclusively on drought resistance metrics (Rt or Rt_{ARIMA}),
249
    and not on the resilience metrics described in Lloret et al. (2011), because (1) we would expect resilience to
    be controlled by a different set of mechanisms, and (2) the findings of (DeSoto et al., 2020) suggest that Rt
251
    is a more important drought response metric for angiosperms in that low resistance to moderate droughts
252
    was a better predictor of mortality during subsequent severe droughts than the resilience metrics.
    Analyses focused on testing the predictions presented in Table 1 with Rt as the response variable, and then
    repeated using Rt_{ARIMA} as the response variable. Models were run for all drought years combined and for
255
    each drought year individually. The general statistical model for hypothesis testing was a mixed effects
256
    model, implemented in the lme4 package in R (Bates et al., 2019). In the multi-year model, we included a
257
    random effect of tree nested within species and a fixed effect of drought year to represent the combined
258
    effects of differences in drought characteristics. Individual year models included a random effect of species.
259
    All models included fixed effects of independent variables of interest (Tables 1,3) as specified below. All
    variables across all best models had variance inflation factors <1.2 (1 + /-0.019). We used AICc to assess
261
    model selection, and conditional/marginal R-squared to assess model fit as implemented in the
262
    AICcmodavg package in R (Mazerolle and portions of code contributed by Dan Linden., 2019). AICc refers
263
    to a corrected version of AICc, and is best suited for small data sizes (see Brewer et al., 2016).
264
    To avoid over-fitting models with five species traits (Table 3) across only 12 species, we did not include all
265
    traits as fixed effects in a single linear mixed model, but rather conducted individual tests of each species
266
    trait to determine the relative importance and appropriateness for inclusion in the main model. These tests
267
    followed the model structure specified above, then added ln[H] and ln[TWI] to create a base model
268
    against which we tested traits. Trait variables were considered appropriate for inclusion in the main model
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    if they had a consistent direction of response across all droughts and if their addition to the base model
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    improved fit (at \triangle AICc \ge 1.0) in at least one drought year (Table S4). We note that we did not use the
271
    \DeltaAICc \geq 1.0 criterion as a test of significance, but rather of whether the variable had enough influence to
    be considered as a candidate variable in full models.
273
    We then determined the top full models for predicting Rt (or Rt_{ARIMA}). To do so, we compared models
274
    with all possible combinations of candidate variables, including ln[H]*ln[TWI] and species traits as
275
    specified above. We identified the full set of models within \triangle AICc=2 of the best model (that with lowest
    AICc). When a variable appeared in all of these models and the sign of the coefficient was consistent across
277
    models, we viewed this as support for the acceptance/rejection of the associated prediction (Table 1). If
278
    the variable appeared in some but not all of these models, and its sign was consistent across models, we
    considered this partial support/rejection. In presentation of the results below, we note instances where the
280
    Rt_{ARIMA} model disagreed with the Rt model, but otherwise do not discuss the Rt_{ARIMA} model.
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    All analysis beyond basic data collection was performed using R version 3.6.2 (R Core Team, 2019). Other
282
    R-packages used in analyses are listed in the Supplementary Information (Appendix S1). All data, code,
283
    and results are available through the SCBI-ForestGEO organization on GitHub
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- 285 (https://github.com/SCBI-ForestGEO: SCBI-ForestGEO-Data and
- 286 McGregor_climate-sensitivity-variation repositories), with static versions corresponding to data and
- ²⁸⁷ analyses presented here archived in Zenodo (DOIs: 10.5281/zenodo.3604993 and [TBD], respectively.

288 Results

- 289 Tree height and microenvironment
- 290 In the years for which we have vertical profiles in climate data (2016-2018), taller trees—or those in
- dominant crown positions—were generally exposed to higher evaporative demand during the peak growing
- 292 season months (May-August; Fig. 2). Specifically, maximum daily wind speeds were significantly higher
- above the top of the canopy (40-50m) than within and below (10-30m) (Fig. 2a). Relative humidity was
- ²⁹⁴ also somewhat lower during June-August, ranging from ~50-80% above the canopy and ~60-90% in the
- understory (Fig. 2b). Air temperature did not vary consistently across the vertical profile (Fig. 2c).
- ²⁹⁶ Crown position varied as expected with height (dominant > co-dominant > intermediate > suppressed),
- but with substantial variation (Fig. 2d). There were significant differences in height across all crown
- position classes (Fig. 2d). A comparison test between height and crown position data from the most recent
- ²⁹⁹ ForestGEO census (2018) revealed a correlation of 0.73.
- 300 Community-level drought responses
- 301 At the community level, cored trees showed substantial growth reductions in all three droughts, with a
- mean Rt of 0.86 in 1966 and 1999, and 0.84 in 1977 (Fig. 2b). Across the entire study period (1950-2009),
- the focal drought years were the three years with the largest fraction of trees exhibiting Rt < 0.7.
- Specifically, in each drought, roughly 30% of the cored trees had growth reductions of > 30% (Rt < 0.7):
- ³⁰⁵ 29% in 1966, 32% in 1977, and 27% in 1999. However, some individuals exhibited increased growth, i.e.,
- Rt > 1.0: 26% of trees in 1966, 22% in 1977, and 26% in 1999.
- In the context of the multivariate model, Rt did not vary across drought years. That is, drought year as a
- variable did not appear in any of the top models -i.e., models that were statistically indistinguishable
- $(\Delta AICc < 2)$ from the best model.
- 310 Tree height, microenvironment, and drought resistance
- Taller trees (based on H in the drought year) showed stronger growth reductions during drought (Table 1;
- Figs. 4, S6). Specifically, ln[H] appeared, with a negative coefficient, in the best model ((Δ AICc=0) and
- all top models when evaluating the three drought years together (Tables S6-S7). The same held true for
- $_{314}$ 1966 individually. For the 1977 drought, ln[H] did not appear in the best model, but was included, with a
- negative coefficient, among the top models-i.e., models that were statistically indistinguishable
- $_{316}$ (\triangle AICc<2) from the best model (Tables 1, S6-S7). For the 1999 drought, ln[H] had no significant effect.
- Rt had a significantly negative response to ln[TWI] across all drought years combined (Figs. 4, S6, Table
- S6-S7). The effect was also significant for 1977 and 1999 individually (Fig. 4, Table S6). When Rt_{ARIMA}
- was used as the response variable, the effect was significant in 1977, and included in some of the top
- models in 1966 and 1999 (Table S7). This negates the idea that trees in moist microsites would be less
- affected by drought. Nevertheless, we tested for a ln[H] * ln[TWI] interaction, a negative sign of which
- could indicate that smaller trees (presumably with smaller rooting volume) are more susceptible to drought

ln[H] * ln[TWI] interaction was never significant, and had a positive sign in any top models in which it appeared (Tables 1, S6-S7). This term did appear with a positive coefficient in the best Rt_{ARIMA} model 325 for all years combined (Table S7), indicating that the negative effect of height on Rt was significantly 326 stronger in wetter microhabitats. Species' traits and drought resistance Species, as a factor in ANOVA, had significant influence (p< 0.05) on all traits (wood density, LMA, 329 PLA_{dry} , and π_{tlp}), with more significant pairwise differences for wood density and PLA_{dry} than for LMA330 and π_{tlp} (Table 2, Fig. S4). Drought resistance also varied across species, overall and in each drought year (Fig. 3). Significant differences in Rt across species were most pronounced in 1966 with a total of seven 332 distinct groupings, while 1977 had four and 1999 had two. Averaged across all droughts, Rt was lowest in 333 Liriodendron tulipifera (mean Rt = 0.66) and highest in Fagus grandifolia (mean Rt = 0.99). Wood density, LMA, and xylem porosity were all poor predictors of Rt (Tables 1,S4-S5). Wood density and LMA were never significantly associated with Rt in the single-variable tests and were therefore 336 excluded from the full models. Xylem porosity was also excluded from the full models, as it had no 337 significant influence for all droughts combined and had contrasting effects in the individual droughts: whereas ring-porous species had higher Rt than diffuse- and semi-ring- porous species in the 1966 and 1999 339 droughts, they had lower Rt in 1977 (Table S4). It is noteworthy that the two diffuse-porous species in our 340 study, Liriodendron tulipifera and Fagus grandifolia, were at opposite ends of the Rt spectrum (Fig. 3), further refuting the idea that xylem porosity is a useful predictor of Rt in the context of this study. 342 In contrast, PLA_{dry} , and π_{tlp} were both negatively correlated to drought resistance (Figs. 4, S6; Tables 343 1,S4-S7). Both had consistent signs across all droughts, and their inclusion at least marginally improved 344 the model ($\triangle AICc > 1.0$) for at least one of the three droughts (Table S4), qualifying them as candidate 345 variables for the full model. PLA_{dry} had a significant influence, with negative coefficient, in full models for 346 the three droughts combined and for the 1966 drought individually (Fig. 4; Tables S6-S7). For 1977 and 347 1999, it was included with a negative coefficient in some of the top models (Tables S6-S7). π_{tlp} was included 348 with a negative coefficient in the best model for both all droughts combined and for the 1977 drought 349 individually (Fig. 4; Table S6). It was also included in some of the top models for 1999 (Tables S6-S7). 350

in drier microenvironments with a deeper water table. This hypothesis was rejected, as the

351 Discussion

Tree height, microenvironment, and leaf drought tolerance traits shaped tree growth responses across three 352 droughts at our study site (Table 1, Fig. 4). The greater susceptibility of larger trees to drought, similar to 353 forests worldwide (Bennett et al., 2015), was driven primarily by their height (Stovall et al., 2019). Taller 354 height was likely a liability in itself, and was also associated with greater exposure to conditions that would 355 promote water loss and heat damage during drought (Fig. 2). There was no evidence that greater 356 availability of, or access to, soil water availability increased drought resistance; in contrast, trees in wetter 357 topographic positions had lower Rt (Zuleta et al., 2017; Stovall et al., 2019), and the larger potential 358 rooting volume of large trees provided no advantage in the drier microenvironments. The negative effect of height on Rt held after accounting for species' traits, which is consistent with recent work finding height 360 had a stronger influence on mortality risk than forest type during drought (Stovall et al., 2020). Drought 361 resistance was not consistently linked to species' LMA, wood density, or xylem type (ring- vs. diffuse

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porous), but was negatively correlated with leaf drought tolerance traits (PLA_{dry}, \pi_{tlp}). This is the first
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    study to our knowledge linking PLA_{dry} and \pi_{tlp} to growth reduction during drought. The directions of
    these responses were consistent across droughts (Table S6), supporting the premise that they were driven
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    by fundamental physiological mechanisms. However, the strengths of each predictor varied across droughts
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    (Fig. 4; Tables S6-S7), indicating that drought characteristics interact with tree size, microenvironment,
    and traits to shape which individuals are most affected. These findings advance our knowledge of the
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    factors that make trees vulnerable to growth declines during drought and, by extension, likely make them
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    more vulnerable to mortality (Sapes et al., 2019).
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    The droughts considered here were of a magnitude that has occurred with an average frequency of
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    approximately once every 10-15 years (Fig. 1a, Helcoski et al., 2019) and had substantial but not
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    devastating impacts on tree growth (Figs. 1b). These droughts were classified as severe (PDSI < -3.0;
    1977) or extreme (PDSI < -4.0; 1966, 1999) at our site and have been linked to tree mortality in the
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    eastern United States (Druckenbrod et al., 2019). However, extreme, multiannual droughts such as the
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    so-called "megadroughts" of this type that have triggered massive tree die-off in other regions (e.g., Allen
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    et al., 2010; Stovall et al., 2019) have not occurred in the Eastern United States within the past several
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    decades (Clark et al., 2016). Of the droughts considered here, the 1966 drought, which was preceded by
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    two years of dry conditions (Fig. S1), severely stressed a larger portion of trees (Fig. 1b). The tendency
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    for large trees to have lowest resistance was most pronounced in this drought, consistent with other
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    findings that this physiological response increases with drought severity (Bennett et al., 2015; Stovall et al.,
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    2019). Across all three droughts, the majority of trees experienced reduced growth, but a substantial
    portion had increased growth (Fig. 1b), potentially due to decreased leaf area of competitors during the
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    drought (REF-if we can find one), and consistent with prior observations that smaller trees can exhibit
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    increased growth rates during drought (Bennett et al., 2015). It is likely because of the moderate impact of
    these droughts, along with other factors influencing tree growth (e.g., stand dynamics), that our best
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    models characterize only a modest amount of variation in Rt: 11-12% for all droughts combined, and
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    18-25% for each individual drought (Fig. S6; Table S6).
    Consistent with studies in other forests worldwide (Bennett et al., 2015), taller trees in this forest exhibited
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    lower drought resistance. Mechanistically, this is consistent with, and reinforces, previous findings that
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    biophysical constraints make it impossible for trees to efficiently transport water to great heights and
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    simultaneously maintain strong resistance and resilience to drought-induced embolism (Olson et al., 2018;
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    Couvreur et al., 2018; Roskilly et al., 2019). Taller trees also face dramatically different microenvironments
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    (Fig. 2). They are exposed to higher wind speeds and lower humidity (Fig. 2a-b), resulting in higher
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    evaporative demand. Unlike other temperate forests where modestly cooler understory conditions have
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    been documented (Zellweger et al., 2019), particularly under drier conditions (Davis et al., 2019), we
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    observed no significant variation in air temperatures across the vertical profile (Fig. 2c). More critically for
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    tree physiology, leaf temperatures can become significantly elevated over air temperature under conditions
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    of high solar radiation and low stomatal conductance (Campbell and Norman, 1998; Rey-Sánchez et al.,
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    2016). Under drought, when air temperatures tend to be warmer, direct solar radiation tends to be higher
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    (because of less cloud cover), and less water is available for evaporative cooling of the leaves, trees with
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    sun-exposed crowns may not be able to simultaneously maintain leaf temperatures below damaging
    extremes and avoid drought-induced embolism. Indeed, previous studies have shown lower drought
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    resistance in more exposed trees (Liu and Muller, 1993; Suarez et al., 2004; Scharnweber et al., 2019).
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    Unfortunately, collinearity between height and crown exposure in this study (Fig. 2d) makes it impossible
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and mature forest stands, along with short and tall isolated trees, would be valuable for more clearly disentangling the roles of tree height and crown exposure. 408 Belowground, taller trees would tend to have larger root systems (Enquist and Niklas, 2002), but this does 409 not necessarily imply that they have greater access to or reliance on deep soil-water resources that may be 410 critical during drought. Rather, larger trees may allocate more to abundant shallow roots that are beneficial for taking up water from rainstorms (Meinzer et al., 1999). In any case, the potentially greater 412 access to water did not override the disadvantage conferred by height-and, in fact, greater moisture access 413 in non-drought years (here, higher TWI) appears to make trees more sensitive to drought (Zuleta et al., 414 2017; Stovall et al., 2019). This may be because moister habitats would tend to support species and 415 individuals with more mesophytic traits (Bartlett et al., 2016b; Mencuccini, 2003; Medeiros et al., 2019), 416 potentially growing to greater heights (e.g., Detto et al. (2013)), and these are then more vulnerable when drought hits. The observed height-sensitivity of Rt, together with the lack of conferred advantage to large 418 stature in drier topographic positions, agrees with the concept that physiological limitations to 419 transpiration under drought shift from soil water availability to the plant-atmosphere interface as forests age (Bretfeld et al., 2018), such that tall, dominant trees are the most sensitive in mature forests. Again, 421 additional research comparing drought responses across forests with different tree heights and water 422 availability would be valuable for disentangling the relative importance of above- and belowground 423 mechanisms across trees of different size. 424 The development of tree-ring chronologies for the twelve most dominant tree species at our site (Helcoski 425 et al., 2019; Bourg et al., 2013) gave us the sample size to compare historical drought responses across 426 species (Fig. 3) and associated traits at a single site (see also Elliott et al., 2015). Our study reinforced 427 current understanding (see Introduction) that wood density and LMA are not reliably linked to drought 428 resistance (Table 1). Contrary to previous studies in temperate deciduous forests, we did not find an 429 association between xylem porosity and drought tolerance, as the two diffuse-porous species, Liriodendron 430 tulipifera and Fagus grandifolia, were at opposite ends of the Rt spectrum (Fig. 3). While the low Rt of L. 431 tulipifera is consistent with other studies (Elliott et al., 2015), the high Rt of F. grandifolia contrasts with 432 studies identifying diffuse porous species in general (Elliott et al., 2015; Kannenberg et al., 2019), and the 433 genus Faqus in particular (Friedrichs et al., 2009), as drought sensitive. There are two potential 434 explanations for this discrepancy. First, other traits can and do override the influence of xylem porosity on drought resistance. Ring-porous species are restricted mainly to temperate deciduous forests (Wheeler 436 et al., 2007), while highly drought-tolerant diffuse-porous species exist in other biomes (REFS). Faqus 437 grandifolia had intermediate π_{tlp} and low PLA_{dry} (Fig. S4), which would have contributed to its drought 438 resistance (Fig. 4; see discussion below). A second explanation of why F. grandifolia trees at this particular 439 site had higher Rt is that the sampled individuals, reflective of the population within the plot, are generally 440 shorter and in less-dominant canopy positions compared to most other species (Fig. S4). The species, which is highly shade-tolerant, also has deep crowns (Anderson-Teixeira et al., 2015b), implying that a 442 lower proportion of leaves would be affected by harsher microclimatic conditions at the top of the canopy 443 under drought (Fig. 2). Thus, the high Rt of the sampled F. grandifolia population can be explained by a combination of fairly drought-resistant leaf traits, shorter stature, and a buffered microenvironment. 445 Concerted measurement of tree-rings and leaf drought tolerance traits of emerging importance (Scoffoni et al., 2014; Bartlett et al., 2016a; Medeiros et al., 2019) allowed novel insights into the role of drought

to confidently partition causality. Additional research comparing drought responses of early successional

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tolerance traits in shaping drought response. The finding that PLA_{dry} and π_{tlp} can be useful for predicting drought responses of tree growth (Fig. 4; Table 1) is both novel and consistent with previous studies linking these traits to habitat and drought tolerance. Previous studies have demonstrated that π_{tlp} and 450 PLA_{dry} are physiologically meaningful traits linked to species distribution along moisture gradients 451 (Maréchaux et al., 2015; Fletcher et al., 2018; Medeiros et al., 2019; Simeone et al., 2019; Rosas et al., 2019; Zhu et al., 2018), and our findings indicate that these traits also influence drought responses. Furthermore, 453 the observed linkage of π_{tlp} to Rt in this forest aligns with observations in the Amazon that π_{tlp} is higher 454 in drought-intolerant than drought-tolerant plant functional type. Further, it adds support to the idea that 455 this trait is useful for categorizing and representing species' drought responses in models (Powell et al., 456 2017). Because both PLA_{dry} and π_{tlp} can be measured relatively easily (Bartlett et al., 2012; Scoffoni 457 et al., 2014), they hold promise for predicting drought growth responses across diverse forests. The importance of predicting drought responses from species traits increases with tree species diversity; whereas 459 it is feasible to study drought responses for all dominant species in most boreal and temperate forests (e.g., 460 this study), this becomes difficult to impossible for species that do not form annual rings, and for diverse 461 tropical forests. Although progress is being made for the tropics (Schöngart et al., 2017), a full linkage of 462 drought tolerance traits to drought responses would be invaluable for forecasting how little-known species 463 and whole forests will respond to future droughts (Christoffersen et al., 2016; Powell et al., 2017). As climate change drives increasing drought in many of the world's forests (Trenberth et al., 2014; 465 Intergovernmental Panel on Climate Change, 2015), the fate of forests and their climate feedbacks will be 466 shaped by the biophysical and physiological drivers observed here. Our results, consistent with other 467 observations around the world, imply that the tallest, most exposed trees will be most affected (Bennett 468 et al., 2015; Stovall et al., 2019). We show that, at least within the mature forest studied here, the 469 vulnerability conferred by tall height and associated crown exposure outweigh any advantage of a larger 470 root system, even in drier microenvironments. This would suggest that the drought responses of trees in 471 mature forests are more strongly differentiated along the size spectrum by their above- than below-ground 472 environment. The same may not be true of systems where short trees exist outside of a buffered understory 473 environment-i.e., open grown trees or short-statured, early-successional forests. The latter appear to be 474 limited more strongly by root water access during drought (Bretfeld et al., 2018), and would also be 475 dominated by species with different traits. The earlier-successional species at our site (Liriodendron tulipifera, Quercus spp., Fraxinus americana) display a mix of traits conferring drought tolerance and 477 resistance (Table 2), while the late-successional Fagus grandifolia displayed high drought resistance, in part 478 because it exists primarily within a buffered microenvironment. Further research on how leaf drought 479 tolerance traits and drought vulnerability change over the course of succession would be valuable for 480 addressing how drought tolerance changes as forests age (e.g. Rodríguez-Catón et al., 2015). In the 481 meantime, the results of this study advance our knowledge of the factors conferring drought resistance in a 482 mature forest, opening the door for more accurate forecasting of forest responses to future drought. 483

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495 Author Contribution

- 496 KAT, IM, and AJT designed the research. Tree-ring chronologies were developed by RH under guidance of
- ⁴⁹⁷ AJT and NP. Trait data were collected by IM, JZ under guidance of NK and LS. Other plot data were
- collected by IM, AS, EGA, and NB under guidance of EGA and WM. Data analyses were performed by IM
- under guidance of KAT and VH. KAT and IM interpreted the results. IM and KAT wrote the first draft of
- manuscript, and all authors contributed to revisions.

501 Supplementary Information

- 502 redo this list!!
- Table S1: Species-specific bark thickness regression equations
- Table S2: Species-specific height regression equations
- Table S3: Palmer drought severity index (PDSI) by month for focal droughts
- Figure S1: Map of ForestGEO plot showing TWI and location of cored trees
- 507 Figure S2: Time series of Palmer Drought Severity Index (PDSI) for the 2.5 years prior to each focal
- 508 drought
- Figure S3: Height (from reconstructed DBH) by crown position across the three focal droughts and in the
- year of measurement (2018)

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